Application of Information Theory, Lecture 11

Pseudo-Entropy and Pseudorandom Generators

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Part I

Motivation

Definition 1

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A pair of algorithms (E, D) is (perfectly correct) encryption scheme, if for any $k \in \{0,1\}^n$ and $m \in \{0,1\}^\ell$, it holds that D(k, E(k,m)) = m

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- Statistical security? HW. Computational security?

Part II

Statistical Vs. Computational distance

Distributions and statistical distance

Let \mathcal{P} and \mathcal{Q} be two distributions over a finite set \mathcal{U} . Their statistical distance (also known as, variation distance) is defined as

$$\mathsf{SD}(\mathcal{P},\mathcal{Q}) := \frac{1}{2} \sum_{x \in \mathcal{U}} |\mathcal{P}(x) - \mathcal{Q}(x)| = \max_{\mathcal{S} \subseteq \mathcal{U}} (\mathcal{P}(\mathcal{S}) - \mathcal{Q}(\mathcal{S}))$$

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For any pair of (finite) distributions \mathcal{P} and \mathcal{Q} , it holds that

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Let $\mathcal{P}, \mathcal{Q}, \mathcal{R}$ be finite distributions, then

Triangle inequality: $SD(P, R) \leq SD(P, Q) + SD(Q, R)$

Repeated sampling: $SD(\mathcal{P}^2 = (\mathcal{P}, \mathcal{P}), \mathcal{Q}^2 = (\mathcal{Q}, \mathcal{Q})) \leq 2 \cdot SD(\mathcal{P}, \mathcal{Q})$

Section 1

Computational Indistinguishability

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- Can it be different from the statistical case?
- ▶ Unless said otherwise, distributions are over $\{0,1\}^n$

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- ▶ Hence, $\varepsilon' < 2\varepsilon$ implies s' > s n.

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Proof: ?

► For $i \in \{0, ..., k\}$, let $H^i = (P_1, ..., P_i, Q_{i+1}, ..., Q_k)$, where the P_i 's are iid $\sim \mathcal{P}$ and the Q_i 's are iid $\sim \mathcal{Q}$. (hybrids)

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Assume \mathcal{P} and \mathcal{Q} are (s, ε) -indistinguishable, then \mathcal{P}^k and \mathcal{Q}^k are $(s - kn, k\varepsilon)$ -indistinguishable.

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- ▶ Thus, $\varepsilon' \le k\varepsilon$ implies s' > s kn
- When considering bounded time algorithms, things behaves very differently!

Part III

Pseudorandom Generators

Definition 6 (pseudorandom distributions)

A distribution \mathcal{P} over $\{0,1\}^n$ is (s,ε) -pseudorandom, if it is (s,ε) -indistinguishable from U_n .

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- Do such generators exist?
- Applications?

Section 2

Pseudorandom generators (PRGs) from One-Way Permutations (OWPs)

Claim 8

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Let $f: \{0,1\}^n \mapsto \{0,1\}^n$ be a poly-time permutation and let $b: \{0,1\}^n \mapsto \{0,1\}$ be a poly-time (s,ε) -hardcore predicate of f, then g(x) = (f(x),b(x)) is a $(s-O(n),\varepsilon)$ -PRG.

▶ Hence, OWP ⇒ PRG

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- ▶ Proof: Let D be an s'-size algorithm with $\Delta^{D}(g(U_n), U_{n+1}) = \varepsilon'$, we will show $\exists (s' + O(n))$ -size P with $\Pr[P(f(U_n)) = b(U_n)] = \frac{1}{2} + \varepsilon'$.

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$$\delta = \Pr[\mathsf{D}(f(U_n), U_1) = 1]$$

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$$= \Pr[U_1 = b(U_n)] \cdot \Pr[D(f(U_n), U_1) = 1 \mid U_1 = b(U_n)]$$

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OWP to PRG

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▶ Hence, $\Pr\left[D(f(U_n), \overline{b(U_n)}) = 1\right] = \delta - \varepsilon'$

- ▶ $Pr[D(f(U_n), b(U_n)) = 1] = \delta + \varepsilon'$
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Algorithm 9 (P)

Input: $y \in \{0, 1\}^n$

- 1. Flip a random coin $c \leftarrow \{0, 1\}$.
- **2.** If D(y, c) = 1 output c, otherwise, output \overline{c} .

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$$\Pr\left[\mathsf{P}(f(U_n)) = b(U_n)\right] = \Pr[c = b(U_n)] \cdot \Pr\left[\mathsf{D}(f(U_n), c) = 1 \mid c = b(U_n)\right] \\ + \Pr[c = \overline{b(U_n)}] \cdot \Pr\left[\mathsf{D}(f(U_n), c) = 0 \mid c = \overline{b(U_n)}\right]$$

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$$\begin{aligned} \Pr[\mathsf{P}(f(U_n)) &= b(U_n)] = \Pr[c = b(U_n)] \cdot \Pr[\mathsf{D}(f(U_n), c) = 1 \mid c = b(U_n)] \\ &+ \Pr[c = \overline{b(U_n)}] \cdot \Pr[\mathsf{D}(f(U_n), c) = 0 \mid c = \overline{b(U_n)}] \\ &= \frac{1}{2} \cdot (\delta + \varepsilon') + \frac{1}{2} (1 - \delta + \varepsilon') = \frac{1}{2} + \varepsilon'. \end{aligned}$$

Part IV

PRG from Regular OWF

Definition 10

X has (s, ε) -pseudoentropy at least k, if \exists rv Y with $H(Y) \ge k$ and $\Delta^{D}(X, Y) \le \varepsilon$ for any s-size D. (s, ε) -pseudo min/Reiny -entropy are analogously defined.

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Example

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- Repeated sampling

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- Example
- Repeated sampling
- Non-monotonicity
- Ensembles
- ▶ In the following we will simply write (s, ε) -entropy, etc

High entropy OWF from regular OWF

Claim 11

Let $f: \{0,1\}^n \mapsto \{0,1\}^n$ be a 2^k -regular (s,ε) -one-way, let $\mathcal{H} = \{h: \{0,1\}^n \mapsto \{0,1\}^{k+2}\}$ be 2-universal family, and let g(h,x) = (f(x),h,h(x)). Then

- **1.** $H_2(g(U_n, H)) \ge 2n \frac{1}{2}$, for $H \leftarrow \mathcal{H}$.
- **2.** g is $(\Theta(s\varepsilon^2), 2\varepsilon)$ -one-way.
- \blacktriangleright k and m and \mathcal{H} are parameterized by of n
- ▶ We assume $\log |\mathcal{H}| = n$ and $s \ge n$

$$\begin{aligned} \mathsf{CP}(g(U_n, H)) &:= \Pr_{w, w' \leftarrow \{0,1\}^n \times \mathcal{H}} [g(w) = g(w')] \\ &= \Pr_{h, h' \leftarrow \mathcal{H}} [h = h'] \cdot \Pr_{(x, x') \leftarrow (\{0,1\}^n)^2} [f(x) = f(x')] \\ &\cdot \Pr_{h \leftarrow \mathcal{H}; (x, x') \leftarrow (\{0,1\}^n)^2} [h(x) = h(x') \mid f(x) = f(x')] \end{aligned}$$

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$$\begin{split} \mathsf{CP}(g(U_n,H)) &:= \Pr_{w,w' \leftarrow \{0,1\}^n \times \mathcal{H}} \left[g(w) = g(w') \right] \\ &= \Pr_{h,h' \leftarrow \mathcal{H}} \left[h = h' \right] \cdot \Pr_{(x,x') \leftarrow (\{0,1\}^n)^2} \left[f(x) = f(x') \right] \\ &\cdot \Pr_{h \leftarrow \mathcal{H}; (x,x') \leftarrow (\{0,1\}^n)^2} \left[h(x) = h(x') \mid f(x) = f(x') \right] \\ &= \mathsf{CP}(\mathsf{H}) \cdot \mathsf{CP}(f(U_n)) \cdot (2^{-k} + (1 - 2^{-k}) \cdot 2^{-k-2}) \\ &\leq \mathsf{CP}(H) \cdot \mathsf{CP}(f(U_n)) \cdot 2^{-k} \cdot \frac{4}{3} = 2^{-n} \cdot 2^{-n} \cdot \frac{4}{3}. \end{split}$$

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Hence, $H_2(g(U_n, H)) \ge 2n + \log \frac{3}{4} \ge 2n - \frac{1}{2}$.

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Consider the following inverter for f

Algorithm 12 (B)

Input: $y \in \{0, 1\}^n$.

Return D(y, h, z), for $h \leftarrow \mathcal{H}$ and $z \leftarrow \{0, 1\}^{\ell}$.

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Input: $y \in \{0,1\}^n$, $h \in \mathcal{H}$ and $z_1 \in \{0,1\}^{\ell}$.

For all $z_2 \in \{0, 1\}^{k+2-\ell}$:

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$$SD((f(x), h, h(x)_{1,...,\ell})_{x \leftarrow \{0,1\}, h \leftarrow \mathcal{H}}, (f(x), h, U_{\ell})_{x \leftarrow \{0,1\}, h \leftarrow \mathcal{H}}) \le \varepsilon'/2$$
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Hence,

$$\Pr_{x \leftarrow \{0,1\}^n} \left[\mathsf{B}(f(x)) \in f^{-1}(f(x)) \right] \ge \varepsilon' - \varepsilon'/2 = \varepsilon'/2.$$

Claim 14

Let $g: \{0,1\}^n \mapsto \{0,1\}^m$ be a function with $H_2(g(U_n)) \ge n - \frac{1}{2}$, and let b be (s,ε) -hardcore predicate for g. Then $v(U_n) = (g(U_n),b(U_n))$ has (s,ε) -Renyi-entropy $n+\frac{1}{2}$.

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The function $v^n(x_1, \ldots, x_n) = (v(x_1), \ldots, v(x_n))$ has $(s - n^2, n\varepsilon)$ -Renyi-entropy $n^2 + \frac{n}{2}$.

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Let $\mathcal{H}: \{0,1\}^{n^2+n} \mapsto \{0,1\}^{n^2+n/4}$ be an 2-universal family and let $G: \{0,1\}^n \times \mathcal{H}$ defined by $G(x_1,\ldots,x_n,h) = (h,h(v^n(x_1,\ldots,x_n)))$. Then $G(H,U_n^n)$ is $(s-n^2-s_{\mathcal{H}},n\varepsilon+2^{-n/4})$ indistinguishable from $(H,U_{n^2+n/4})$, for $H \leftarrow \mathcal{H}$ and $s_{\mathcal{H}}$ being the size of sampling and evaluating algorithm for \mathcal{H} .

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If f and b and \mathcal{H} (?) are poly-time computable, then G is a $(s-n^2-s_{\mathcal{H}},n_{\mathcal{E}}+2^{-n/4})$ -PRG.

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